

# Package ‘SmoothTensor’

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**Type** Package

**Title** A Collection of Smooth Tensor Estimation Methods

**Version** 0.1.1

**Imports** methods, Matrix, rTensor

**Description** A list of methods for estimating a smooth tensor with an unknown permutation. It also contains several multi-variate functions for generating permuted signal tensors and corresponding observed tensors. For a detailed introduction for the model and estimation techniques, see the paper by Chanwoo Lee and Miaoyan Wang (2021) “Smooth tensor estimation with unknown permutations” <[arXiv:2111.04681](https://arxiv.org/abs/2111.04681)>.

**URL** <https://arxiv.org/abs/2111.04681>

**License** GPL-3

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Borda_count	<i>Borda count algorithm for nonparametric tensor estimation with unknown permutation.</i>
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**Description**

Estimate a signal tensor and permutation from a noisy and incomplete data tensor using Borda count estimation method.

**Usage**

```
Borda_count(A, l, kvec, sym = FALSE)
```

**Arguments**

A	A given (possibly noisy and incomplete) data tensor. Missing value should be encoded as NA.
l	Degree of polynomial approximation.
kvec	A vector of the number of groups for each mode.
sym	Boolean variables representing symmetricity of the signal tensor. Non-symmetric tensor (sym = FALSE) is default.

**Value**

The returned object is a list of components.

theta - An estimated signal tensor based on Borda count estimation.

permutation - An estimated permutation based on Borda count estimation.

**References**

C. Lee and M. Wang. Smooth tensor estimation with unknown permutations. arXiv:2111.04681, 2021.

**Examples**

```
# Generate the noisy observation from smooth tensor and permutation
d = 20
sim1 = simulation(d, mode = 1)
signal_T = sim1$signal
observe_T = sim1$observe
permutation = sim1$permutation

# Estimate signal tensor and permutation
kvec = c(3,3,3)
result = Borda_count(observe_T, 2, kvec, sym = TRUE)
```

```
# Calculate MSE
hatTheta = result$Theta
mean((hatTheta-signal_T)^2)
```

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LSE *The least squares estimation for nonparametric tensor estimation with unknown permutation.*

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### Description

Estimate a permuted signal tensor from a noisy data tensor based on the least squares estimation with constant block approximation.

### Usage

```
LSE(A, kvec, sym = FALSE, mode = 3)
```

### Arguments

A	A given noisy data tensor.
kvec	A vector of the number of groups for each mode.
sym	Boolean variables representing symmetricity of the signal tensor. Non-symmetric tensor (sym = FALSE) is default.
mode	An integer from 1 to 3 representing a type of methods for estimating the clustering functions. Higher-order spectral clustering method is default. mode = 1: k-means algorithm applied on unfolded matrices. mode = 2: k-means algorithm for community detection in stochastic block model (only available on binary observation). mode = 3: higher-order spectral clustering algorithm.

### Value

An estimated permuted signal tensor based on the least squares estimation.

### References

C. Gao, Y. Lu, and H. H. Zhou. Rate-optimal graphon estimation. *The Annals of Statistics*, 2015.

K. Balasubramanian. Nonparametric modeling of higher-order interactions via hypergraphons. *Journal of Machine Learning Research*, 2021.

R. Han, Y. Luo, M. Wang, and A. R. Zhang. Exact clustering in tensor block model: Statistical optimality and computational limit. arXiv:2012.09996, 2020.

## Examples

```
# Generate the noisy observation from smooth tensor and permutation
d = 20
sim1 = simulation(d, mode = 1)
signal_T = sim1$signal
observe_T = sim1$observe
permutation = sim1$permutation
psignal_T = signal_T[permutation,permutation,permutation]

# Estimate permuted signal tensor
kvec = c(10,10,10)
hatpTheta = LSE(observe_T,kvec,sym = TRUE)

# Calculate MSE
mean((hatpTheta-psignal_T)^2)
```

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ltns

*Chicago crime tensor dataset*

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## Description

Chicago crime dataset consists of crime counts reported in the city of Chicago, ranging from January 1st, 2001 to December 11th, 2017.

## Usage

```
ltns
```

## Format

An order-3 tensor with entries representing the log counts of crimes from 24 hours, 77 community areas, and 32 crime types.

## Source

<http://frostdt.io/tensors/chicago-crime/>

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mode_info	<i>A list of mode information of the Chicago crime tensor dataset</i>
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**Description**

A list of mode information of order-3 tensor dataset 1tns.

**Usage**

```
mode_info
```

**Format**

A list consisting of crime areas, crime hours, and crime types:

hour\_map 24 hours of crimes

area\_map 77 areas of crimes

crimetype\_map 32 types of crimes

**Source**

<http://frostdt.io/tensors/chicago-crime/>

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simulation	<i>Generate a symmetric tensor observation from the smooth signal tensor, Gaussian noise tensor, and permutation.</i>
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**Description**

Generate a symmetric tensor observation from the smooth signal tensor, Gaussian noise tensor, and permutation. Users can select one of 5 different smooth signal tensors generated from functions specified in Table 4 of the reference given below.

**Usage**

```
simulation(d, mode = 1, sigma = 0.5, signal_level=5)
```

**Arguments**

d	Dimension of a tensor to be generated.
mode	An integer from 1 to 5 corresponding to models specified. Default model is 1.
sigma	Standard deviation of the Gaussian noise tensor. Default value is 0.5.
signal_level	A scale of the magnitude of the signal tensor to be generated.

**Value**

The returned object is a list of components.

signal - A true signal tensor generated from a function specified.

observe - A noisy observation generated from the smooth signal tensor, Gaussian noise tensor, and permutation.

permutation - A true permutation.

**References**

C. Lee and M. Wang. Smooth tensor estimation with unknown permutations. arXiv:2111.04681, 2021.

**Examples**

```
d = 20
# Generate 20 by 20 by 20 observed tensor generated from model 1
sim1 = simulation(d, mode = 1)
observed_tensor = sim1$observe
signal_tensor = sim1$signal
permutation = sim1$permutation
```

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simulation_asym	<i>Generate a non-symmetric tensor observation from the smooth signal tensor, Gaussian noise tensor, and permutation.</i>
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**Description**

Generate a non-symmetric tensor observation from the smooth signal tensor, Gaussian noise tensor, and permutation. Users can select one of 5 different smooth signal tensors generated from functions specified in Table 5 of the reference given below.

**Usage**

```
simulation_asym(d, mode = 1, sigma = 0.5, signal_level=5)
```

**Arguments**

d	A vector of dimensions of a tensor to be generated.
mode	An integer from 1 to 5 corresponding to models specified. Default model is 1.
sigma	Standard deviation of the Gaussian noise tensor. Default value is 0.5.
signal_level	A scale of the magnitude of the signal tensor to be generated.

**Value**

The returned object is a list of components.

signal - A true non-symmetric signal tensor generated from a function specified.

observe - A noisy observation generated from the smooth signal tensor, Gaussian noise tensor, and permutation.

permutation - A list of true permutation for each mode.

**References**

C. Lee and M. Wang. Smooth tensor estimation with unknown permutations. arXiv:2111.04681, 2021.

**Examples**

```
d = c(10,20,30)
# Generate 10 by 20 by 30 observed tensor generated from model 1
sim1 = simulation_asym(d,mode = 1)
observed_tensor = sim1$observe
signal_tensor = sim1$signal
permutation = sim1$permutation
```

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simulation_bin	<i>Generate a symmetric binary tensor from the probability tensor and permutation.</i>
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**Description**

Generate a symmetric binary tensor from the probability tensor and permutation. Users can select one of 5 different smooth probability tensor generated from functions specified in Table 4 of the reference given below.

**Usage**

```
simulation_bin(d, mode = 1)
```

**Arguments**

d	Dimension of a tensor to be generated.
mode	An integer from 1 to 5 corresponding to models specified. Default model is 1.

**Value**

The returned object is a list of components.

signal - A true probability tensor generated from a function specified.

observe - A binary tensor generated by Bernoulli trials given the probability tensor and permutation.

permutation - A true permutation.

**References**

C. Lee and M. Wang. Smooth tensor estimation with unknown permutations. arXiv:2111.04681, 2021.

**Examples**

```
d = 20
# Generate 20 by 20 by 20 binary-valued tensor generated from model 1
sim1 = simulation_bin(d, mode = 1)
observed_tensor = sim1$observe
signal_tensor = sim1$signal
permutation = sim1$permutation
```

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Spectral	<i>Spectral method for nonparametric tensor estimation with unknown permutation.</i>
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**Description**

Estimate a permuted signal tensor from a noisy data tensor using spectral method, which performs universal singular value thresholding on the unfolded tensor.

**Usage**

```
Spectral(A, row_idx, col_idx, threshold = NULL)
```

**Arguments**

A	A given noisy data tensor.
row_idx	The indices of the modes that map onto the row space
col_idx	The indices of the modes that map onto the column space
threshold	A threshold to disregard singular values. Default value is the square root of unfolded matrix dimension.

**Value**

An estimated permuted signal tensor based on Spectral method.

**References**

J. Xu. Rates of convergence of spectral methods for graphon estimation. International Conference on Machine Learning, 2018.  
 C. Lee and M. Wang. Smooth tensor estimation with unknown permutations. arXiv:2111.04681, 2021.



**Examples**

```
# Generate the noisy observation from smooth tensor and permutation
d = 20
sim1 = simulation(d,mode = 1)
signal_T = sim1$signal
observe_T = sim1$observe
permutation = sim1$permutation
psignal_T = signal_T[permutation,permutation,permutation]

# Estimate permuted signal tensor
hatpTheta = Spectral(observe_T,1,c(2,3))

# Calculate MSE
mean((hatpTheta-psignal_T)^2)
```

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